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Selectivity of the First Two Glycerol Dehydrogenation Steps Determined Using Scaling Relationships

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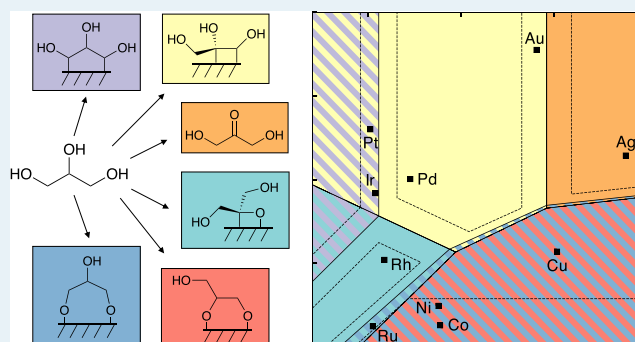
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ABSTRACT: Glycerol is a byproduct of biodiesel production and an abundant feedstock that can be used for the synthesis of high-value chemicals. There are many approaches for glycerol valorization, but, due to the complicated reaction mechanism, controlling which products are produced is challenging. Here, we describe glycerol's chemical selectivity for different metallic catalysts using descriptors for carbon (mainly *C , *CH_2OH) and oxygen (mainly *O , CH_3O^*). The quality of these descriptors and the weighted combinations thereof are validated based on their fit, via linear regression, to the binding energies of all reaction intermediates generated in the first two glycerol dehydrogenation steps on a number of close-packed Ru, Co, Rh, Ir, Ni, Pd, Pt, Cu, Ag, and Au surfaces. We show that *CH_2OH is a better descriptor than *C for the studied carbon-bound intermediates, which is attributed to the observation that the adjacent *OH group interacts with the surface. This leads to a negative oxygen dependence, which can be generalized to similar alcohol-derived adsorbates. Furthermore, we show that CH_3O^* is a better oxygen descriptor than *O for the studied intermediates. This is mainly attributed to the difference between the single and double bonds, as we show that *OH is closer to the accuracy of CH_3O^* . Multilinear regression with different combinations of *C , *O , and *OH is comparable in accuracy to that of *CH_2OH and CH_3O^* . Scaling relationships are used to determine the selectivity map for glycerol dehydrogenation. The results show that the first dehydrogenation is selective toward two different intermediates (one bonded via the secondary carbon and the other via the secondary oxygen) depending on the relative bond strength of the carbon and oxygen descriptors. The second dehydrogenation step results in five intermediates, again depending primarily on the relative bond strength of carbon and oxygen to the surface. The selectivity maps can be used together with kinetic considerations and experimental data to find catalyst candidates for glycerol dehydrogenation.

KEYWORDS: scaling, glycerol, CH_2OH , CH_3O , first principles, selectivity, transition metals



INTRODUCTION

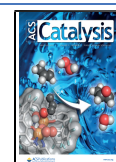
Over the years, it has become clear that developing renewable and clean energy resources will be essential in the coming decades. In this context, biofuels produced from vegetable oils and waste fats are an alternative to petroleum-based fuels.^{1,2} One concern, due to large-scale production of biofuels, is the massive amount of glycerol that is formed as a byproduct. It is estimated that the biodiesel industry generates hundreds of millions of kilograms of low-value glycerol,^{3–7} and because of its abundance, it is important to find a practical utilization to support the production of biodiesel. One alternative is to use glycerol to produce hydrogen gas for fuel cell applications^{8–10} or to reduce iron ores in steel production.^{11,12} An advantage of glycerol is the fact that the C–C–C backbone is relatively stable, which makes it possible to interrupt the oxidative reaction to form valuable chemical products without producing CO_2 .^{4,8} Another benefit is that hydrogen generation using glycerol as a starting product can be performed at a lower potential compared to using water and other alcohols.^{10,13–15}

In the past few decades, numerous studies have focused on glycerol valorization and how to produce useful products.^{7,9,10,16} One of the methods that has shown promise is glycerol electrooxidation (GEO).¹⁷ Products of the GEO process include tartronic acid (TA), dihydroxyacetone (DHA), glyceraldehyde (GLYD), glyceric acid (GLYC), hydroxypyruvic acid (HYDP), mesoxalic acid (MESOX), and glycolic acid (GA). All of these are three-carbon products of GEO and they have broad applications. They can be used as food additives,⁹ in agriculture,¹⁸ in detergent production,⁹ in biodegradable polymer synthesis,⁶ as tanning agents in cosmetics,¹⁹ and in medical applications.^{20,21}

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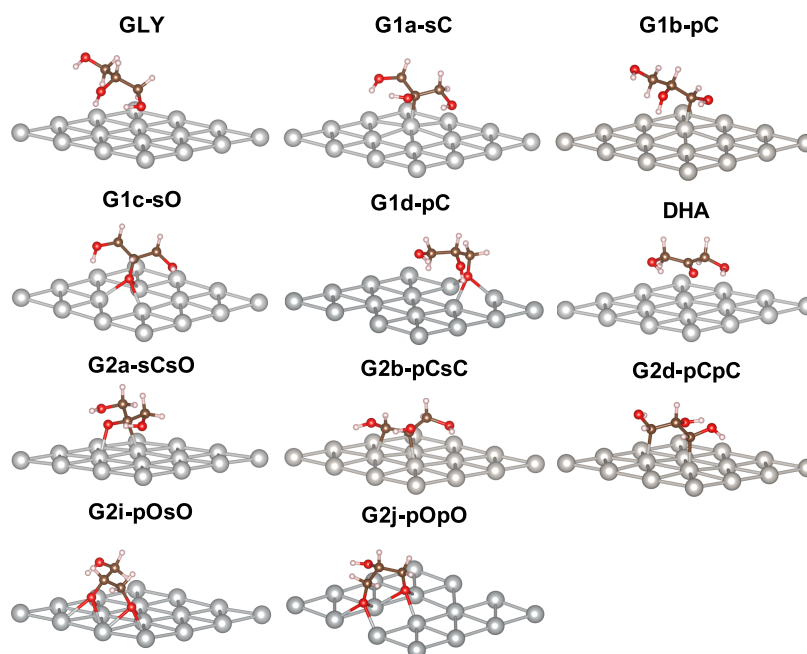


Figure 1. Glycerol intermediate structures. For each intermediate, the optimal adsorption site for most of the metals is shown. The specific adsorption site for each metal is indicated in Table 1.

As mentioned above, the dehydrogenation of glycerol is a complex chemical process, even when compared with other alcohols. Glycerol is a large molecule, and its possible intermediary steps and products are almost innumerable.⁸ Particularly from an atomistic modeling perspective, it is desirable to determine the energetically preferred species on the surface of the catalyst and find all possible elementary steps that could lead to the products observed experimentally. It involves a series of reaction steps involving C–H and O–H bond breaking (at higher potentials C–C bond breaking is also observed) and, hence, its success depends heavily on the chemical affinity of the catalyst. A material that interacts strongly with an adsorbed carbon atom (*C) tends to induce more C–H bond splitting, while on the other hand, a catalyst that interacts strongly with an oxygen atom (*O) will tend to induce O–H bond splitting.

It is known that the adsorption energies for *O and *C display linear scaling correlations with the reaction intermediates in many reactions.^{22–27} In this sense, theoretical calculations can be used to classify different materials based on their chemical affinity and thus establish a microscopic description of the reactivity of the catalyst.^{28,29} In addition, it has been found that the activation energy scales linearly with the final-state adsorption energy, that is Brønsted–Evans–Polanyi (BEP)-type relationships.²⁶ Together, adsorption energy scaling and BEP relationships lead to the concept of volcano curves in catalysis, which is a useful tool to find new catalyst formulations.^{28,30–32}

Scaling relations were used by Liu and Greeley³³ to estimate the binding energies of glycerol intermediates on a number of transition metals based on the binding energy on platinum, assuming that it is proportional to the valency of the adsorbate, as originally proposed by Abild-Pedersen et al.²² In comparison to DFT calculations, the thus obtained energies for palladium and rhodium agreed quite well, but larger deviations were reported for copper. This was attributed to the copper surface

being unable to break double bonds, in which case the bond-counting valency scheme breaks down.³³

In the present work, we establish scaling relations for glycerol intermediates with respect to the binding energies of atomic carbon (*C) and oxygen (*O) as descriptors without any a priori assumption of valency. In addition, we introduce two new descriptors, hydroxymethyl (*CH₂OH) and methoxy (CH₃O*), which have a better fit to the binding energies of the reaction intermediates and provide a better distinction between different metals. We extensively discuss the reasons for the difference in accuracy of the pairs of descriptors and how this understanding may aid in general predictions of alcohol intermediates. After investigating the sensitivity on the exchange-correlation functional, the two pairs of descriptors are used to generate selectivity maps for the first two steps in glycerol dehydrogenation, which can be used to search for possible catalysts for the reaction.

METHOD

Density functional calculations were performed using VASP,^{34–36} where the projector augmented wave method³⁷ was used to model the interaction between the valence electrons and the core. The Kohn–Sham orbitals were represented using a plane-wave basis set with 450 eV as cutoff energy and a Gaussian smearing of 0.05 eV was applied to the Fermi-level discontinuity. Calculations on Ni and Co were performed using spin polarization.

The main exchange-correlation functional applied in the present work, which is used unless otherwise specified, was optB86b-vdW, which includes the non-local effects of electron correlation.^{38–40} optB86b-vdW is known to describe the adsorption of molecules and layered materials accurately.^{41,42} The computational setup used herein was verified in a previous study of methanol adsorption on Au.⁴³ Furthermore, the BEEF-vdW functional⁴⁴ was used to determine how sensitive the results are with respect to the exchange-correlation

Table 1. Optimal Adsorption Sites for Descriptors and Intermediates

species	Ag	Au	Co	Cu	Ir	Ni	Pd	Pt	Rh	Ru
C	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc
O	fcc	fcc	hcp	fcc	fcc	fcc	fcc	fcc	fcc	hcp
OH	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc
CH ₂ OH	atop	atop	bridge	bridge	atop	bridge	atop	atop	atop	bridge
CH ₃ O	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc	fcc
GLY	atop	atop	atop	atop	atop	atop	atop	atop	atop	atop
G1a-sC	atop	atop	atop	atop	atop	atop	atop	atop	atop	atop
G1b-pC	atop	atop	atop	atop	atop	atop	atop	atop	atop	atop
G1c-sO	fcc	fcc	fcc	fcc	fcc	fcc	fcc	atop	fcc	fcc
G1d-pO	fcc	fcc	fcc	fcc	atop	fcc	fcc	atop	fcc	fcc
DHA	vdW	vdW		vdW	vdW	vdW	vdW	vdW	vdW	
G2a-sCsO	atop	atop	atop	atop	atop	atop	atop	atop	atop	atop
	atop	atop	atop	atop	atop	atop	atop	atop	atop	bridge
G2b-pCsC	vdW	vdW	fcc	vdW	atop	atop	atop	atop	atop	atop
			atop		atop	atop	atop	atop	atop	atop
G2d-pCpC	atop	atop	atop	atop	atop	atop	atop	atop	atop	atop
	atop	atop	atop	atop	atop	atop	atop	atop	atop	atop
G2i-pOsO	fcc	bridge	fcc	fcc	atop	fcc	fcc	atop	fcc	fcc
	fcc	atop	fcc	fcc	atop	fcc	bridge	bridge	fcc	fcc
G2j-pOpO	fcc	fcc	fcc	fcc	atop	fcc	fcc	atop	fcc	fcc
	fcc	fcc	fcc	fcc	atop	fcc	fcc	atop	fcc	fcc

functional. The reason for comparing optB86b/vdWDF and BEEF for the same system is that recent studies indicate that these two functionals are on opposite sides when comparing the over- and under-binding of various molecules, e.g., benzene, naphthalene, on Pt.^{45,46} Finally, the PBE functional⁴⁷ was used for some reference calculations on scaling relations and on van der Waals effects.

The close-packed surfaces—hcp(0001) for Ru and Co, and fcc(111) for the rest—were modeled as 4-layer p(3 × 3) supercells, sampled in a Monkhorst–Pack grid⁴⁸ with (6 × 6 × 1) *k*-points. The periodic surface slabs were separated by a vacuum of 20 Å. Gas-phase species were computed in a (20 × 20 × 20) Å³ cell using only the Γ point. Structures were relaxed to a maximum residual force of 0.02 eV/Å. The reported adsorbate energies correspond to the optimized adsorption sites, unless otherwise specified.

The binding energy for a species is defined as the energy of the adsorbed species relative to the bare surface and gas-phase reference and the stoichiometric amount of hydrogen gas. The gas-phase reference was glycerol for glycerol derivatives, methanol for *CH₂OH and CH₃O*, (spin-polarized) oxygen gas for *O, and (spin-polarized) atomic carbon for *C. As the compared structures are similar, in particular with respect to vibrational modes, entropy contributions and zero-point corrections are considered to be small and are not taken into account.

We perform no direct calculations on barriers, neither in a heterogeneous catalysis framework nor in an electrochemical environment. The work is focused on thermodynamics, and barriers are discussed based on previous works.

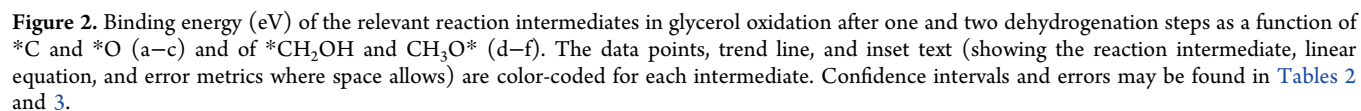
Scaling relations are established by linear regression, for which the accuracy and reliability are analyzed by calculating 95% confidence intervals of the slopes, mean absolute errors (MAE), maximum absolute errors (MAX), and coefficient of determination (*R*²). It should be noted that MAE and MAX are better error metrics than *R*² in this work, as *R*² lacks robustness for small sample sizes,⁴⁹ which is why the analysis is

mostly based on MAE and MAX along with the confidence interval.

RESULTS AND DISCUSSION

In a recent study,⁵⁰ the structure and energy were calculated for all reaction intermediates in glycerol oxidation after the first two deprotonation steps on the close-packed surfaces of cobalt, nickel, copper, ruthenium, rhodium, palladium, silver, iridium, platinum, and gold using DFT. The most stable intermediates and their structure can be seen in Figure 1. For each intermediate, the optimal adsorption site for most metals is shown; the specific adsorption for each metal is indicated in Table 1. Each intermediate is labeled according to the origin (glycerol; G), the number of deprotonations (0, 1, 2), and an additional running index (a–j) to distinguish intermediates, followed by the carbon and oxygen atoms formally binding to the surface (p = primary, s = secondary). Glycerol and dihydroxyacetone are only denoted with abbreviations (GLY and DHA), as they form no formal bonds with the surface but rather interact in a van der Waals fashion. Furthermore, for G2b-pCsC, three bonding mechanisms can be distinguished: The coinage metals do not break the double bond (C–C bond length ~ 1.35 Å); Co, Ni, and Ru partially break the double bond (C–C bond length ~ 1.45 Å) by binding the carbon atoms to one surface atom; Pd, Pt, Rh, and Ir break the double bond (C–C bond length ≥ 1.5 Å) by binding the carbon atoms to separate surface atoms.

Establishing Scaling Relations for Glycerol Intermediates. Using linear regression, we start by fitting the traditional descriptors, the binding energy of carbon (*C) and oxygen (*O), bound at optimal hollow sites, to the computed binding energy of the intermediates as shown in Figure 2a–c, under the assumption that the bond is either carbon-like or oxygen-like (with the exception of G2a-sCsO that has formally one bond of each kind). Since *C has a valency of four and *O a valency of two, we expect single carbon-surface bonds to contribute a slope of 0.25 and single oxygen-surface bonds to



intermediate	*C slope	MAE (eV)	MAX (eV)	R ²	*CH ₂ OH Slope	MAE (eV)	MAX (eV)	R ²
GLY	0.13 ± 0.02	0.03	0.06	0.96	0.33 ± 0.08	0.04	0.09	0.92
G1a-sC	0.40 ± 0.12	0.14	0.39	0.87	1.11 ± 0.25	0.12	0.26	0.93
G1b-pC	0.40 ± 0.06	0.08	0.19	0.97	1.07 ± 0.11	0.05	0.12	0.98
G2b-pCsC	0.35 ± 0.11	0.15	0.31	0.87	0.98 ± 0.20	0.09	0.21	0.94
G2d-pCpC	0.72 ± 0.18	0.25	0.39	0.91	2.00 ± 0.19	0.09	0.16	0.99
DHA	0.09 ± 0.02	0.02	0.06	0.94	0.23 ± 0.09	0.04	0.07	0.86

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Table 3. *O and CH₃O* Scaling Relations^a

intermediate	*O slope	MAE (eV)	MAX (eV)	R ²	CH ₃ O* slope	MAE (eV)	MAX (eV)	R ²
G1c-sO	0.65 ± 0.15	0.14	0.30	0.92	1.13 ± 0.12	0.06	0.16	0.98
G1d-pO	0.59 ± 0.11	0.09	0.23	0.95	1.00 ± 0.17	0.08	0.22	0.96
G2i-pOsO	1.04 ± 0.28	0.27	0.55	0.90	1.83 ± 0.15	0.08	0.17	0.99
G2j-pOpO	1.11 ± 0.30	0.27	0.63	0.90	1.95 ± 0.14	0.08	0.17	0.99

^aConfidence intervals and error estimates.

contribute 0.5.²² Confidence intervals on the slopes along with other error estimates are displayed in Tables 2 and 3.

As seen in Figure 2a and Table 2, glycerol and DHA have small but clearly positive slopes, which is surprising as these species have no formal surface bond. With the mean and maximum absolute errors being less than or equal to 0.03 and 0.06 eV, respectively, and the coefficient of determination being 0.94 or higher, the relations are quite accurate. Moving on to the formally carbon-bound species, we expect G1a-sC and G1b-pC to have one carbon bond and G2d-pCpC to have two carbon bonds, while G2b-pCsC could have anything between zero and two based on whether the C–C double bond is broken or not, as discussed above. The slope for both G1a-sC and G1b-pC is 0.40 and that for G2d-pCpC is 0.72; these are all greater than the formally expected 0.25 and 0.5, which do not fall within the confidence intervals. G2b-pCsC has a slope of 0.35 ± 0.11, which suggests a single bond, based both on the results for G1a-sC and G1b-pC and on the fact that 0.25 falls within the interval. The errors are relatively large, in particular for G1a-sC and G2d-pCpC, with MAE being 0.14 and 0.25 eV, respectively, and MAX being almost 0.4 eV. There is an agglomeration of *C binding energies for the non-coinage metals and a large separation for the coinage metals. While this behavior is in line with the literature,^{22,51} the predictive power is low as it is hard to distinguish between the non-coinage metals. In the case of G1a-sC and G2d-pCpC, in particular, there is a steeper trend among these metals reflected in the large maximum absolute errors.

For the singly oxygen-bonded intermediates (Figure 2b and Table 3), we get a slope of 0.65 ± 0.15 and 0.59 ± 0.11 for G1c-sO and G1d-pO, respectively, with the expected 0.5 from a single oxygen bond at the lower ends of the intervals. The values for the intermediates with two formal bonds, G2i and G2j (1.04 ± 0.28 and 1.11 ± 0.30), have the expected 1.0 well within the intervals. However, the errors are also in this case large, in particular for G2i-pOsO and G2j-pOpO (MAE of 0.27 eV and MAX of 0.55 and 0.63 eV). The reason for this can be seen in Figure 2b, as copper and silver follow a trend below the fitted line, while platinum, palladium, and iridium follow another trend above it.

Through multilinear regression, G2a-sCsO is fitted to a weighted combination of *C and *O (Figure 2c). G2a-sCsO binds through one carbon and one oxygen, which formally should give coefficients 0.25 and 0.5 for *C and *O, respectively. As we find the coefficients to be 0.44 ± 0.38 and 0.41 ± 0.44, with an MAE of 0.19 eV, MAX of 0.29 eV, and R² of 0.78, it is clear that the fit is not very accurate. At most, it can be said that it appears as though the carbon bond contributes more than the oxygen bond in this representation.

To investigate the issues with the fits to *C and *O, we fit the binding energies to two alternative descriptors, *CH₂OH and CH₃O* (Figure 2d–f and Tables 2 and 3), based on the assumption that the similar alcohol structure will make them more similar to glycerol intermediates, with a better balance

between carbon-like and oxygen-like bonds. Since these descriptors can each form one formal bond, the slopes should simply be the number of bonds of the relevant intermediate (which can form more than one bond) according to the valency model. The calculated slopes for the singly deprotonated species are in the range 1.00–1.13, with the expected 1.0 within or just outside the confidence interval, while G2d-pCpC, G2i-pOsO, and G2j-pOpO are in the range 1.83–2.00, with the expected 2.0 within or just outside the confidence interval. G2b-pCsC has a slope of 0.98 ± 0.20, which we again interpret as a single bond. For G2a-sCsO, the expected slope is 1.0 for each descriptor, while the calculated values are 1.31 ± 0.57 and 0.63 ± 0.44 for *CH₂OH and CH₃O*, respectively; compared to *C and *O, the errors are lower across the board, with the exception of GLY and DHA. Furthermore, there is a greater spread in terms of descriptor binding energies (x-axis) when *CH₂OH is used as a descriptor. The latter aspect allows more robust discrimination among the non-coinage metals.

Descriptor Analysis. Why are *CH₂OH and CH₃O* better descriptors for glycerol intermediate binding energies than *O and *C? One thing to test is the effect of the matching adsorption sites, as Abild-Pedersen et al.²² showed that matching the adsorption site improves the scaling relations. Since the goal of the present work is to predict the optimal binding energy of glycerol intermediates, without knowing the optimal adsorption site on each metal (or alloy) surface a priori, the only degree of freedom in this regard is to fix the adsorption site of the descriptor. Another possible test is to use a descriptor with a single bond, similar to the intermediates and to CH₃O*.

In Table 4, we fit CH₃O* to *O (in an optimized adsorption site), *O atop, and *OH (in an optimized adsorption site). It

Table 4. Scaling Relations with CH₃O*^a

O descriptor	slope vs CH ₃ O*	MAE (eV)	MAX (eV)	R ²
*O	0.56 ± 0.16	0.14	0.30	0.89
*O(atop)	0.54 ± 0.25	0.21	0.49	0.75
*OH	0.99 ± 0.10	0.06	0.14	0.98

^aConfidence intervals and error estimates.

turns out that *OH is the best fit to CH₃O*. We conclude then that having an oxygen descriptor with one surface bond is the most important in this case. If *OH is used as a descriptor for the oxygen-bound intermediates, it still is not as accurate as CH₃O* (see Supporting Information). This can possibly be attributed to the hyperconjugation in the C–O bond.⁵²

All the carbon-bound intermediates prefer binding atop, while *CH₂OH binds atop on most metals and in a bridge position on others. In order to investigate the importance of having similar adsorption sites, we fit the intermediates to *C atop in Figure 3 and Table 5. It can be noted that the *C binding energies are more spread out for the non-coinage

metals, but the errors are comparable to those of the optimized *C (Table 2).

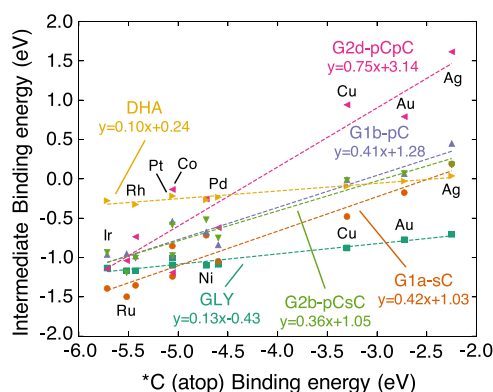


Figure 3. Binding energy (eV) of the relevant reaction intermediates in glycerol oxidation after one and two dehydrogenation steps as a function of *C atop. The data points, trend line, and inset text (showing the reaction intermediate and linear equation) are color-coded for each intermediate. Confidence intervals and errors can be found in Table 5. In comparison to Figure 2a, the *C binding energies are no longer clustered for the non-coinage metals.

Table 5. *C(atop) Scaling Relations^a

intermediate	*C(atop) slope	MAE	MAX	R ²
*CH ₂ OH	0.38 ± 0.11	0.12	0.20	0.91
GLY	0.13 ± 0.03	0.04	0.09	0.93
G1a-sC	0.42 ± 0.11	0.13	0.24	0.92
G1b-pC	0.41 ± 0.11	0.12	0.26	0.92
G2b-pCsC	0.36 ± 0.10	0.11	0.20	0.92
G2d-pCpC	0.75 ± 0.24	0.27	0.56	0.88
DHA	0.10 ± 0.02	0.03	0.04	0.94

^aConfidence intervals and error estimates.

Similar to the oxygen case, we tested to establish a scaling relation between *CH₃ and *CH₂OH as they both have one formal surface bond. With a slope of 1.17 ± 0.42 , an MAE of 0.16 eV, and a MAX of 0.26 eV, the fit is, however, not more precise. With *CH₃ atop, the slope is 1.37 ± 0.29 , MAE 0.10 eV, and MAX 0.22 eV, which is comparable to the values for *C in an optimal adsorption site. It can be noted that the slope should be 1.0, which further indicates that something else is missing.

Since neither scaling *C atop with the atop-bound intermediates nor having the carbon descriptor with a single bond, like the intermediates, resolves the discrepancy between *C and *CH₂OH, we investigate whether the oxygen also interacts with the surface. Such interaction is indicated by the fact that among the non-coinage metals, the oxygen-surface distance varies from 2.25 Å on ruthenium to 3.15 Å on

platinum, two metals with the same *C binding energy but very different *O binding energy. Thus, the intermediate surface bonds would not be described as merely either carbon-like or oxygen-like.

The multilinear regression analysis of *CH₂OH is shown in Table 6. It can be seen that the dependence on the oxygen descriptor tends to be slightly negative. In the case of the pairs *C, *O and *C, *OH, an oxygen-associated slope of zero is within or at the edge of the confidence interval; in isolation, these two results would not imply a statistically significant correlation with a negative slope. However, with *C atop, the slope of the oxygen descriptor is more clearly negative; this shows that *C atop is a more accurate descriptor for the carbon contribution of the bond. The multilinear analysis was replicated with the PBE exchange-correlation functional (see Supporting Information) with similar results on *C. Furthermore, *OH and CH₃O*, i.e., the more accurate oxygen descriptors based on the results presented above, have clearer negative oxygen-associated slopes. These results imply that the quality of the descriptors is important to observe the negative oxygen dependence.

The negative oxygen dependence can be tentatively explained based on the oxygen electronegativity; if the surface interacts weakly with oxygen, charge is taken from the carbon atom, resulting in a less satisfied valency and, as a consequence, stronger carbon interaction with the surface to compensate. On the other hand, if the surface interacts strongly with the oxygen, the metal will provide the charge directly; thus, the carbon has less to gain by interacting with the surface and binds more weakly. Supporting this is the observation that the Bader charge on the carbon atom is $\sim 0.3e$ lower on Pt, Pd, Ag, and Au with relatively weak oxygen-surface interaction compared with Co, Ni, Cu, and Ru with stronger oxygen preference, while Rh and Ir fall halfway between. It can be noted that Co, Ni, Cu, and Ru bind *CH₂OH on a bridge site, such that a C–H bond is stretched, which has been observed for similar systems as a stabilizing donation–backdonation mechanism.^{53,54} However, if only atop configurations (preferred for the rest of the metals) are compared, the charge difference persists, though smaller ($\sim 0.2e$). The C–O bond length is more stretched for the oxygen-preferring group compared with the carbon-preferring one; as an example, on platinum the bond length is 1.39 Å, while on cobalt it is elongated to 1.47 Å, because in the former case the bond is more polar covalent. Finally, a fixed carbon affinity but increasing oxygen affinity results in weaker adsorption; cobalt and platinum have the same carbon binding energy atop while all carbon-bound intermediates except GLY and DHA bind stronger on platinum; in particular, it can be seen in Figure 3, e.g., that G2d-pCpC binds 1.0 eV stronger on platinum.

Turning to GLY and DHA, we have seen above that they scale well with *C and *CH₂OH. As they visually seem to

Table 6. Scaling Relations with *CH₂OH^a

C descriptor	slope vs *CH ₂ OH	O descriptor	slope vs *CH ₂ OH	MAE (eV)	MAX (eV)	R ²
*C	0.41 ± 0.09	*O	−0.10 ± 0.13	0.07	0.18	0.96
*C	0.39 ± 0.06	*OH	−0.15 ± 0.15	0.07	0.17	0.99
*C(atop)	0.52 ± 0.10	*O(atop)	−0.24 ± 0.15	0.06	0.15	0.97
*C(atop)	0.42 ± 0.07	*OH	−0.22 ± 0.16	0.07	0.18	0.97
C(atop)	0.43 ± 0.08	CH ₃ O	−0.23 ± 0.17	0.07	0.18	0.97

^aConfidence intervals and error estimates.

interact with a surface atom through their primary OH group (see Figure 1), it might be worthwhile to investigate their scaling with oxygen descriptors atop. *O atop vs GLY has a slope of 0.15 ± 0.10 , an MAE of 0.08 eV, and a MAX of 0.15 eV, while DHA has a slope of 0.15 ± 0.08 , an MAE of 0.05 eV, and a MAX of 0.08 eV. CH_3O^* atop vs GLY has a slope of 0.33 ± 0.17 , an MAE of 0.07 eV, and a MAX of 0.12 eV, while DHA has a slope of 0.30 ± 0.15 , an MAE of 0.05 eV, and a MAX of 0.08 eV. So, while the margins of the error are larger and the errors are higher than shown in Table 2, the positive slopes are statistically significant. Similar results have been reproduced using the PBE exchange functional, which lacks vdW interaction (see Supporting Information). Thus, some caution should be taken in concluding the correlation to be carbon-like. On the other hand, *C seems to be better at representing the interaction, which can be shown by fitting GLY to *C and *O atop, with the clearly positive *C slope of 0.11 ± 0.03 and unclear *O (atop) slope of 0.02 ± 0.03 .

* CH_2OH and CH_3O^* are the most accurate descriptors for all non-vdW intermediates. By multilinear regression, it can be shown that only one of them at a time scales with a given intermediate, with a slope of 0 for the other descriptor within the confidence interval, in addition to not improving the MAE (see Supporting Information). The multilinear regression of combinations of *C, *O, and *OH in different adsorption sites shows that the carbon-binding intermediates have a negative dependence on *OH, which in most cases is non-significant due to the relatively large margins of error (see Supporting Information). It is still reasonable to believe that the oxygen dependence for * CH_2OH plays some role in these intermediates based on the adsorption geometry and the fits to * CH_2OH . The advantage of the multilinear regression of *C, *O, and *OH is that there are extensive databases on these descriptors,⁵⁵ which can be used to extrapolate to other surfaces.

Exchange-Correlation Functional Sensitivity. It has been shown that the use of optB86b-vdW gives good agreement with the experiment regarding adsorption on coinage metals^{41,43} and, through explicit inclusion of vdW, the interaction between layered materials.⁴² However, it is still important to estimate the sensitivity of the present results to the choice of the functional. In a recent study of the molecular adsorption on Pt,⁴⁶ optB86b-vdW and BEEF-vdW were often found to overbind and underbind, respectively, when compared to experimental data. Here, the BEEF-vdW functional was used to evaluate the sensitivity of our results. Figure 4 shows * CH_2OH and CH_3O^* evaluated with optB86b-vdW and BEEF-vdW. In terms of slope, the descriptors have almost one-to-one correlation (slope of 0.98 ± 0.23 and 0.92 ± 0.20 , respectively) with MAE being 0.08 and 0.05 eV and MAX being 0.14 and 0.09 eV. BEEF-vdW underbinds both descriptors relative to optB86b-vdW (0.4 and 0.3 eV for * CH_2OH and CH_3O^* , respectively). By this, we conclude that the differences in terms of slope are comparable to the errors in the regression analysis and, consequently, that the choice of xc-functional should not matter for the conclusions regarding the descriptor analysis above.

Selectivity Maps. Based on the simple scaling relations established for the paired descriptors *C, *O and * CH_2OH , CH_3O^* in Figure 2, we construct thermodynamic selectivity maps for the products of the first and second dehydrogenation steps of glycerol oxidation, which are shown in Figure 5. It should be pointed out that the borders between the regions are

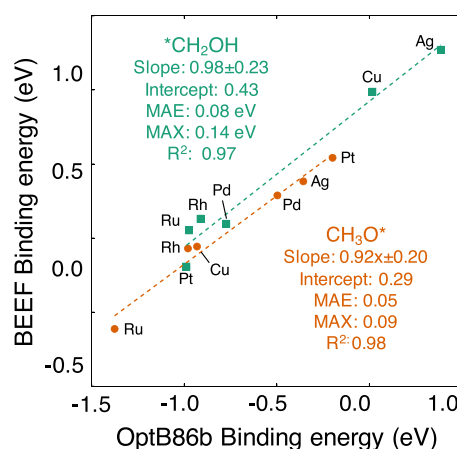


Figure 4. Binding energies of the first dehydrogenation step, evaluated using two different exchange-correlation functionals, optB86b-vdW and BEEF-vdW, both including non-local correlation. Energies are in eV.

not sharp; the selectivity does not go from 100% product A to 100% product B upon crossing. Close to the borders, we assume that there is a Boltzmann-distributed mixture of A and B as a function of the energy difference $\Delta E = |E_A - E_B|$. Thus, the energy margins are shown as dashed lines to indicate sensitivity. The lines are arbitrarily drawn at $\Delta E = 0.1$ eV difference for clarity, but due to linearity, it can be easily visually extrapolated to other margins.

For the first dehydrogenation step (Figure 5a,c), both pairs of descriptors show that the secondary-carbon-bonded G1a-sC and the secondary-oxygen-bonded G1c-sO are most stable. We note that there is an agglomeration of elements for the *C descriptor. For both pairs of descriptors, the map regions are consistent with the actual, directly computed intermediate binding energies, correctly showing the right intermediate for each metal (this is an indicator of the accuracy of the fit). However, with *C and *O, the energy margins are smaller.

For the second dehydrogenation step (Figure 2b,d), we see that the maps are qualitatively similar. As a general rule, it can be seen that the thermodynamically most stable intermediates typically go from two carbon bonds through one carbon bond and one oxygen bond to two oxygen bonds as the relative bond strength of the oxygen descriptor is increased. We distinguish between second-dehydrogenation products available by dehydrogenation of G1a-sC and G1c-sO (bound through the secondary group) on one hand, and G1b-pC and G1d-pO (bound through a primary group) on the other, as seen in Figure 1. Since G1a-sC and G1c-sO are most stable, the products originated by dehydrogenating them (G2a-sCsO, G2b-pCsC, and G2i-pOsO) are referred to as direct products below, while the products stemming from G1b-pC and G1d-pO (G2d-pCpC and G2j-pOpO) are referred to as indirect products. "Indirect" should be interpreted in the sense that their precursors are not the most thermodynamically favored in the first dehydrogenation step. G1b-pC, G1d-pO, G2d-pCpC, and G2j-pOpO could still be produced, however, if the selectivity is less than complete or if the kinetics is more favorable to the dehydrogenation of the primary group.

Starting with the direct products, indicated with solid colors, G2b-pCsC is the preferred intermediate over a wide range from strong to medium bond strength of the carbon descriptor as long as the oxygen descriptor binds somewhat weakly to the

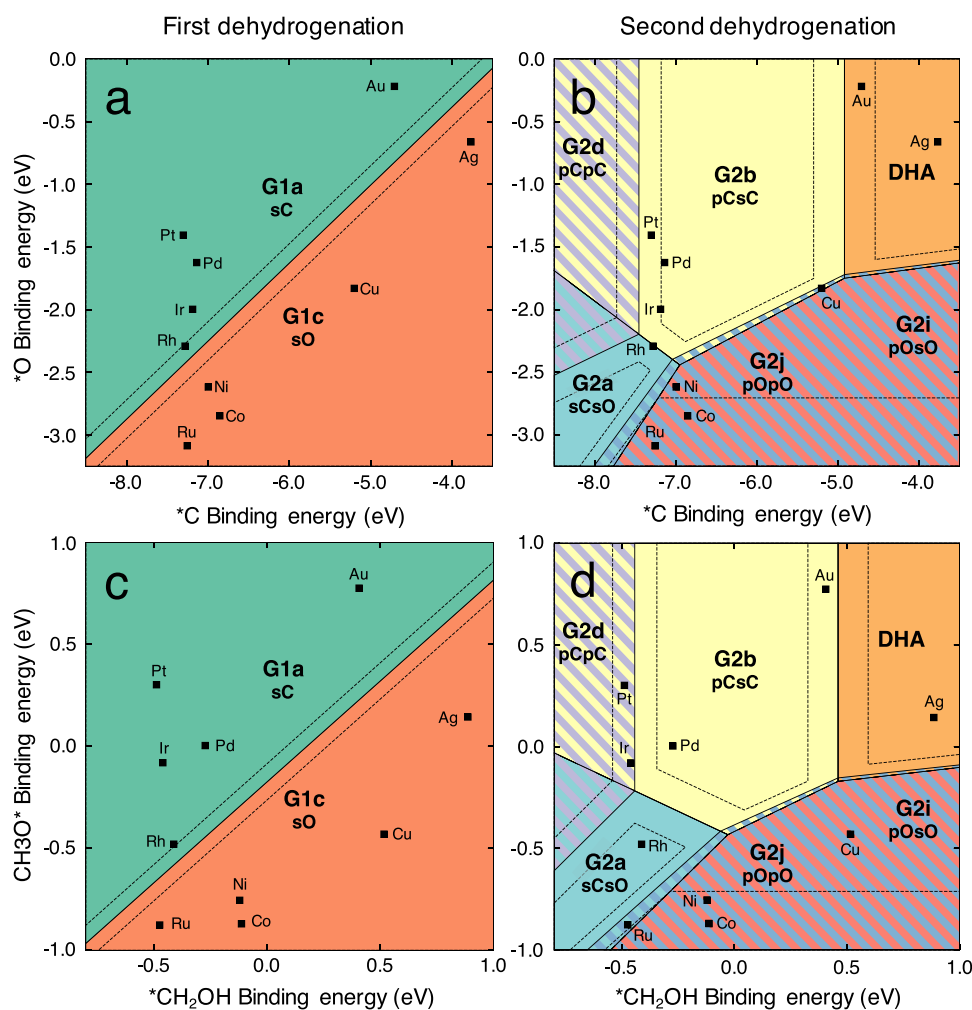


Figure 5. Selectivity for first-dehydrogenated and second-dehydrogenated intermediates of glycerol as a function of the binding energy of O^* and C^* (a, b) and of CH_2OH^* and CH_3O^* (c, d). In the striped areas, the indirect products (see the text for definition) G2d (purple) and G2j (blue) are most stable. Dashed lines indicate 0.1 eV margin for the intermediates.

surface. If both descriptors bind weakly, the van der Waals-bound DHA becomes the preferred product. On the other hand, if both descriptors are strongly bound, G2a-sCsO (DHA with broken $\text{C}=\text{O}$ double bond) is preferred, and if the binding energy of carbon is medium to small, the G2i-pOsO instead dominates. If we include indirect products, there are regimes where G2d-pCpC and G2j-pOpO are the most energetically favorable, preferring strong carbon bonding and oxygen bonding, respectively. These are indicated with striped colors.

For the second dehydrogenation, the maps are mostly consistent, such that the descriptor energies on a metal correctly predict the thermodynamically most stable intermediates from direct DFT calculations. For the C^* and O^* in Figure 5b, the exceptions are (1) iridium, which should favor G2i-pOsO as a direct product and G2d-pCpC as an indirect product; (2) platinum, which should prefer G2d-pCpC as a direct product; and (3) copper, which should form G2i-pOsO as a direct product. For CH_2OH^* and CH_3O^* (Figure 5d), the exceptions are (1) ruthenium, which should exclusively favor G2a-sCsO; (2) iridium, which should prefer G2a-sCsO as a direct product (but G2d-pCpC as indirect); and (3) gold, which should form DHA. All deviations are relatively close to the borders, as seen by the 0.1 eV margin lines. Please note

that G2a-sCsO, in particular, is not too well fitted, as seen above. Furthermore, G2i-pOsO and G2j-pOpO have almost the same energies and slopes, which is reflected by the horizontal dashed line in the lower right region, marking the border if the energy of G2i-pOsO shifts -0.1 eV relative to G2j-pOpO.

So far, we have only discussed thermodynamics, but it is worthwhile to make some remarks on the kinetics. In the presence of water, $\text{O}-\text{H}$ bond breaking is favored over $\text{C}-\text{H}$ bond breaking since water molecules mediate the proton transfer.^{32,56} Even when the reactants are in gas phase, $\text{O}-\text{H}$ bond breaking should be favored over $\text{C}-\text{H}$ bond breaking, relative to other alcohols, because of self-mediation in the form of hydrogen bonding between the three OH groups.^{57,58} Specifically, on rhodium, the barrier is 0.1 eV smaller for $\text{O}-\text{H}$ vs $\text{C}-\text{H}$ bond breaking in glycerol (0.67 vs 0.77 eV) even without the presence of water.⁵⁷ On platinum, the $\text{C}-\text{H}$ breaking is slightly favored (0.71 vs 0.74 eV), while dehydrogenation of the secondary group has a 0.13 eV lower barrier than the primary group.³³ There are, to our knowledge, no studies on glycerol electrooxidation barriers; the qualitative trends are probably similar to the heterogeneous catalysis case, but the magnitude of the relevant barriers has not been determined.

As discussed in our previous paper,⁵⁰ the most common glycerol electrooxidation pathway reported in the experimental literature (typically on platinum, palladium, and gold) is through glyceraldehyde (i.e., deprotonation of the primary carbon and oxygen).⁹ This is in contrast to the calculated surface thermodynamics. The possible causes for this are (1) spontaneous deprotonation of a primary OH group in alkaline solution prior to adsorption;^{59–61} (2) kinetics; (3) steric hindrance; and (4) transformation of the desorbed products in solution, such as Lobry de Bruyn-Alberda van Ekenstein isomerizations between DHA, propene-1,2,3-triol, and glyceraldehyde,^{62,63} since the products are normally detected in solution rather than on the surface. If we also consider non-electrochemical oxidation in water, gold- and platinum-based catalysts are also selective toward glyceraldehyde and its oxidation products,⁷ which excludes the first cause. It is clear that there remains work to be done to connect the theory and experiment in the field.

CONCLUSIONS

The pair of descriptors *C and *O on one hand and *CH_2OH and CH_3O^* on the other are used to establish scaling relations over a number of close-packed transition metals (Ru, Co, Rh, Ir, Ni, Pd, Pt, Cu, Ag, and Au) for the first two steps of glycerol dehydrogenation. We find that the latter pair is better at capturing the behavior in terms of errors, distinguishability with regard to descriptor energy, and intuitive slopes. We find that the main reason for the accuracy of *CH_2OH over *C is that the glycerol intermediates bound with carbon also interact through the adjacent *OH group. For the oxygen descriptors, we find that the difference between CH_3O^* and *O is mostly due to the difference between a single and a double bond, as *OH is comparable in accuracy to CH_3O^* . Exceptions to this trend are the van der Waals-bound glycerol and dihydroxyacetone, which scale well with *C although similar absolute errors are observed with *O .

Using the scaling relations, we find that the first dehydrogenation is selective to oxidation of the carbon or oxygen of the secondary group, depending on the relative binding energy of the descriptors. Continuing from these intermediates in a second dehydrogenation step, we find that the selectivity is steered toward four different intermediates. These are (1) G2b-pCsC, which prefers weak binding energy of the oxygen descriptor and medium to strong binding energy of the carbon descriptor; (2) DHA, which prefers weak binding energy of both descriptors; (3) G2a-sCsO, which prefers strong binding energy of both descriptors; and (4) G2i-pOsO, which prefers stronger binding of the oxygen descriptor than the carbon descriptor. If we consider second-step intermediates stemming from oxidation of the primary group, also G2d-pCpC and G2j-pOpO are most stable in some regions.

The simple scaling relations with *CH_2OH and CH_3O^* can be used to screen for catalyst formulations for glycerol valorization by dehydrogenation. Building on the work in our previous paper,⁵⁰ the selectivity for products such as DHA, GLYD, and HYDP can be deduced from the two first steps of glycerol dehydrogenation. The pathways of continued dehydrogenation on the surface can be studied by establishing new scaling relations to further intermediates that are similar to these two descriptors.

The present work also gives important insight into the nature of any alcohol intermediates bound through carbon

with a hydroxyl group, as the negative dependence on *OH should be common. The alternative to calculating the *CH_2OH (and CH_3O^*) binding energies on a number of surfaces is to use the existing databases of *C , *O , *OH , etc. It is then necessary to expect the oxygen dependence and analyze it through multilinear regression as carried out in the Supporting Information.

ASSOCIATED CONTENT

Supporting Information

The Supporting Information is available free of charge at <https://pubs.acs.org/doi/10.1021/acscatal.0c04186>.

Selectivity of the first two glycerol dehydrogenation steps determined using scaling relationships, containing DFT energies, Bader charges, tabulated scaling relations, and selectivity maps (PDF)

Structure and data files (ZIP)

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Notes

The authors declare no competing financial interest.

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